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ABSTRACT

A concept map is a graph in which the nodes represent concepts, the lines between the nodes represent relations, and the labels on the lines represent the nature of the relations. Concept maps have been used to assess students' knowledge structures, especially in science education. Two concept mapping techniques, constructing a map and filling in a map that has been started were compared to see if the mapping techniques can be considered equivalent, whether the fill-in-the-map techniques are sensitive to the nodes selected to be completed, and whether the fill-in-the-map scores are sensitive to the linking lines selected to be filled-in. Participants were 152 high school chemistry students in 7 classes taught by 2 teachers. On three occasions students constructed or filled in maps as directed. The fill-in (skeleton) map scores were not sensitive to the sample of nodes or linking lines to be filled in. Fill-in-the-nodes and fill-in-the-lines are not equivalent forms of fill-in-the-map, but further research is needed to determine which of these forms provides more accurate information. Results suggest that both mapping techniques are tapping somewhat similar, but not identical, aspects of students' understanding. Construct-a-map scores more accurately reflect the differences across students' knowledge structures, and the relationship between scores from the multiple-choice test and both mapping techniques confirms that the mapping techniques are not equivalent. (Contains 6 tables and 15 references.) (SLD)

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Comparison of the Reliability and Validity of Scores From Two Concept-Mapping Techniques*

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COMPARISON OF THE RELIABILITY AND VALIDITY OF SCORES FROM TWO CONCEPT-MAPPING TECHNIQUES

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Concept maps have been used to assess students' knowledge structures, especially in science education. The justification for assessing student's knowledge structures is based on theory and research showing that understanding a subject domain such as science is associated with a rich set of relations among important concepts in the domain. We know, for example, that successful learners develop elaborate and highly integrated frameworks of related concepts (Mintzes, Wandersee, & Novak, 1997), just as experts do (Chi, Glaser, & Farr, 1988; Glaser, 1991). Furthermore, we know that highly organized structures facilitate problem solving and other cognitive activities (e.g., generating explanations or recognizing rapidly meaningful patterns; Baxter, Elder, Glaser, 1996; Mintzes, Wandersee, & Novak, 1997). Research has shown that differences in the performance of experts and novices are due, largely, to how knowledge is structured in their memories (Chi, Glaser, & Farr, 1988; Glaser, 1991).

Concept maps are interpreted as providing a "picture" of how key concepts in a domain are mentally organized/structured by students. With this assessment technique, students are asked to link pairs of concepts in a science domain and label the links with a brief explanation of how the two concepts go together.

Although concept maps have been used in large-scale, as well as classroom assessments, a wide variety of techniques are called concept maps and little is know about the reliability and validity of scores produced by these assessment techniques (e.g., Ruiz-Primo & Shavelson, 1996). We suspect that the observed characteristics of the representation of a student's knowledge structure depends to a large extent on how the representation is elicited. Simply put, the method used to ask students to represent their knowledge can affect the representation they provide as well as the score they obtain (Ruiz-

Primo & Shavelson, 1996; Ruiz-Primo, Schultz, & Shavelson, 1996; Ruiz-Primo, Shavelson, & Schultz, 1997). Through a series of studies we seek to increase our understanding of how different mapping techniques affect the representation and interpretation of a student's knowledge structure. In this paper, we provide reliability and validity evidence on the effects of two mapping techniques, "*fill-in-the-map*" and "*construct-a-map*."

Concept-Map Assessment

We define a concept map as a graph in which the nodes represent concepts, the lines between nodes represent relations, and the labels on the lines represent the nature of the relations. The combination of two nodes and a labeled line is called a proposition--the fundamental unit of the map. Our characterization of a concept map assessment as based on its three components--a task, its response format, and a scoring system--has revealed the enormity of variations in mapping techniques used in research and practice (see Ruiz-Primo & Shavelson, 1996).

The characteristics of the task, the response format, and the scoring system hold the key for tapping what concept-map based assessments are intended to evaluate: knowledge structure (or "connected understanding" for some authors). The assessment task, for example, can vary in the constraints (directedness) it imposes on a student in eliciting her representation of structural knowledge. One dimension in which directedness varies lies in what is provided for use in the concept map (Figure 1).

	High ← Degree of Directedness →	Low
Concepts	Provided by Assessor	Provided by Student
Linking Lines	Provided by Assessor	Provided by Student
Linking Words	Provided by Assessor	Provided by Student
Structure of the Map	Provided by Assessor	Provided by Student

Figure 1. Degree of directedness in the concept assessment task.

If the characteristics of the assessment task fall on the left extreme the student's representation is probably determined more by the mapping technique (or the assessor if you will), than by the student's own knowledge or connected understanding.[1] If the assessment task falls on the right extreme, the student is free to decide which and how many concepts to include in her map, which concepts are related, and which words to use for explaining the relation. This openness may also be undesirable because of practical issues. For example, asking the student to generate the concepts to construct her map provides a good piece of information about the student's knowledge in a particular domain (e.g., Are the concepts selected by the student relevant/essential to the topic?). However, scoring issues may make this option impractical--for example, each concept map has a unique scoring system. In one of our studies (Ruiz-Primo, Schultz, & Shavelson, 1996) we compared two mapping techniques that differed on whether the concept sample was student-generated or assessor-generated. The student-generated sample technique presented more challenges to score students' representations.

The cognitive demands imposed on students by high-directedness techniques are different from low-directedness techniques. Furthermore, high-directed techniques are more likely to misrepresent the student's knowledge structure by imposing a structure on their responses. In this study we examined the reliability and validity of two mapping techniques, one that can be considered as high-directed and the other as low-directed.

Defining The Two Mapping Techniques

Some researchers (e.g., Schau & Mattern, 1997) have argued that asking students to draw a map from scratch imposes too high a cognitive demand on students to produce a meaningful representation of their knowledge. An alternative technique is the fill-in-the-map. In what follows we describe both techniques, the fill-in-the-map and the construct-a-map.

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[1] The characteristics of the assessment task have an impact on the response format and the scoring system. For example, a task that provides the structure of the map, will probably provide such a structure in the student's response format. If the task provides the concepts to be used, the scoring system will not focus on the "appropriateness of the concepts" used in a map. The combination of the task, the response format, and the scoring system is what determines a mapping technique.

Fill-in the Map. The "fill-in-the-map" technique provides students with a concept map where some of the concepts and/or the linking words have been left out. Students fill-in the blank nodes or blank linking lines (e.g., Anderson & Huang, 1989; McClure & Bell, 1990; Schau, Mattern, Weber, Minnick, & Witt, 1997). The response format is straightforward; students fill-in the blanks and their responses are scored correct-incorrect. Arguments can be made for (e.g., ease of administration, scoring, and retrieval of propositions from long-term memory) and against (e.g., imposes a structure on a student's knowledge) the technique. We posit that as students' subject matter knowledge increases, the structure of their maps should increasingly reflect the structure of the domain as held by experts (see Glaser, in press; Shavelson, 1972, 1974). By imposing a structure on the relations between concepts, it is difficult to know whether or not students' knowledge structures are becoming increasingly similar to experts'. Structure of representation, however, is not the only issue to consider. With "fill-in," students are usually provided with linking words in the skeleton map and they only select the concepts from a list of concepts. Yet, in our research using the construct-a-map technique we found that the linking words students used to relate two concepts provide insight into student's understanding in a particular content domain (e.g., Ruiz-Primo, Schultz, & Shavelson, 1996).

Construct-A-Map From Scratch. The "construct-a-map" technique varies as to how much information is provided by the assessor (Figure 1). The assessor may provide the concepts and/or linking words or may ask students to construct a hierarchical or non-hierarchical map. The response format is simply a piece of paper provided for students to construct the map. Scoring systems vary from counting the number of nodes and linking lines (not recommended) to evaluating the accuracy of propositions (see Ruiz-Primo & Shavelson, 1996).

This mapping technique, however, has been considered problematic for large-scale assessment because students need to be trained to use maps and scoring is difficult and time-consuming (e.g., Schau et al., 1997). Our research has tried to overcome these two problems (see Ruiz-Primo, Schultz & Shavelson, 1996; Ruiz-Primo, Shavelson, & Schultz, 1997). We designed a 50-minute program to teach students how to construct concept maps. The program proved to be effective in achieving this goal with more than 100 high

school students. Moreover, to find an efficient scoring system we have explored different types of scores; some based only on the propositions, others using a criterion map. Map propositions can be scored based on accuracy and comprehensiveness or simply on whether the proposition is correct or incorrect. Based on this differentiation we have studied three types of scores: *proposition accuracy score*--the sum of individual proposition scores obtained on a student's map; *convergence score*--the proportion of accurate propositions in the student's map out of all possible propositions in the criterion map; *salience score*--the proportion of correct propositions out of all propositions in the student's map. (The scoring system we have used has yielded high interrater reliability coefficients, above .90, even when the quality of the propositions is judged.)

Purpose

This study explored the technical characteristics of the "fill-in-the-map" and "construct-a-map" techniques. More specifically, we examined whether the: (a) two mapping techniques can be considered equivalent, (b) fill-in-the-map scores are sensitive to the nodes (concepts) selected to be filled-in (construct-a-map scores have proven not to be sensitive to the sample of concepts used; Ruiz-Primo, Schultz, & Shavelson, 1996), and (c) fill-in-the-map scores are sensitive to the linking lines selected to be filled-in (linking words).

Method

Participants. One hundred and fifty two high school chemistry students and two chemistry teachers participated in the study. Students were in one of seven chemistry classes. Four of the classes were considered advanced; the remainder (56 students) were regular chemistry classes. Two of the four advanced classes were taught by Teacher 1 (six years of teaching experience) and the other two by Teacher 2 (one year of teaching experience). The three regular classes were taught by Teacher 1. All participants were drawn from the Palo Alto area.

Students and teachers were trained to construct concept maps, including the fill-in-the-map technique, with the same 50-minute training program used in previous studies (see Ruiz-Primo, Schultz, & Shavelson, 1996; Ruiz-Primo, Shavelson, & Schultz, 1997). To evaluate the training, 25 percent of the maps

constructed by students at the end of the training session were randomly sampled and analyzed. The analysis focused on whether students used the concepts provided on the list, labeled the lines, and provided accurate propositions. Results indicated that 92 percent of the students used all the concepts provided in the list; and all used labeled lines; and all provided four or more accurate propositions. We concluded that the program succeeded in teaching students to construct concept maps.

Design. To evaluate whether the fill-in the map scores are sensitive to the sample of nodes or linking lines to be filled-in, we used a 2 x 2, concept sample by linking-line sample design. Four 20-node skeleton maps were constructed. In two of the maps 12 nodes (60% of the nodes) were left blank. In the other two skeleton maps, 12 linking lines (31.5% of the linking lines in the criterion map) were left blank (i.e., no linking words). Concepts and linking lines to be left blank were randomly selected from the list of key concepts and the list of propositions in a criterion map. The four skeleton maps were as follows: A--skeleton map with Sample 1 of nodes left blank; B--skeleton map with Sample 2 of nodes left blank; C--skeleton map with Sample 1 of linking lines left blank; and D--skeleton map with Sample 2 of linking lines left blank.

Students were tested on three occasions. On Occasion 1, all students constructed a concept map from scratch using all 20 concepts provided by the assessor. On Occasion 2, half the students filled-in skeleton map A and half filled-in skeleton map B. On Occasion 3 half the students filled-in skeleton map C and half filled-in skeleton map D.

Within each of the 7 classes (groups) students were randomly assigned to one of four sequences of skeleton maps: Sequence 1--skeleton map A followed by skeleton map C; Sequence 2--skeleton map A followed by skeleton map D; Sequence 3--skeleton map B followed by skeleton map C; and Sequence 4--skeleton map B followed by skeleton map D.

Selection of Concepts and Development of the Criterion/Skeleton Map. To identify the structure of the skeleton map for the fill-in mapping technique, we assumed that: (1) there is some "agreed-upon organization" that best reflects the structure of a content domain, (2) "experts" in that domain (in this context, teachers) have a high degree of agreement, and (3) experts' concept maps

provide a reasonable representation of the subject domain (e.g., Glaser, in press). Therefore, the skeleton maps used were based on the criterion map.

We used the topic, "Chemical Names and Formulas," as the domain for sampling the concepts used in the study.[2] Teachers and researchers (the second author was a high school chemistry teacher for 10 years) were involved in the process of selecting the concepts and creating the criterion map. Teachers were asked to identify the concepts they considered to be the most important in the unit. Researchers also selected the most important concepts by carefully reviewing the text used to teach the topic. Figure 2 describes briefly the procedure followed to select the concepts and to define the criterion map (see Ruiz-Primo, Schultz, & Shavelson, 1996).

The "agreed" upon links across teacher's and researcher's maps were represented in the criterion map and considered the "substantial" links that students were expected to know after instruction on the topic. The criterion map was used as the master map for the purpose of constructing the four skeleton maps. The concepts selected for the blank nodes on the skeleton maps were randomly sampled from the key-concept list. The linking lines selected to be filled-in on the skeleton maps were sampled from the linking lines on the criterion map. The propositions provided in the skeleton maps were taken from the criterion map. The concepts for the construct-a-map technique were all those on the key-concept list.

Procedure Used To Construct A Criterion Map	
1.	Ask each participant to provide a list of the 20 most important concepts in the subject domain.
2.	Have participants compare and discuss their lists of selected concepts until a consensus is reached about which are the most important concepts. This will be considered the "Key-Concept List."
3.	Ask each participant to construct a concept map with the key concepts.
4.	Construct a concept map with relations that appear in at least 80% of the participants' concept maps.
5.	Discuss and modify the resulting concept map with participants until a consensus is reached about which relations should be present in the map.
6.	Use the resulting concept map as the "Criterion Map."

Figure 2. Procedure followed to define the Key-Concept List and the Criterion Map.

[2] Although we used this topic in previous studies, the selection of concepts for mapping was carried out again since different teachers participated on this occasion.

Instrumentation. The two mapping techniques varied in their task demands and constraints imposed on students. Figure 3 provides a profile of the directedness of the assessment tasks for both techniques. The construct-a-map technique asked students to construct a map using the 20 concepts provided by the assessor. Students were encouraged to provide propositions (linking words) as specific as they wanted in order to explain the relationship between the two concepts they were linking. No restriction was imposed on the type of structure students could use in the map (e.g., students were not instructed to create a hierarchical structure).

Technique	Concepts	Linking Lines	Linking Words	Structure of the Map
Construct-a-map	Provided	Not-Provided	Not-Provided	Not-Provided
Fill-in-the-map	Provided	Provided	Provided	Provided

Figure 3. Directedness profile of two mapping techniques: *Construct-a-map* and *Fill-in-the-map* .

The fill-in-the-map technique asked students to fill-in two skeleton maps, one with blank nodes and the other with blank linking lines. After randomly selecting nodes only six nodes were different between skeleton map A and skeleton map B. For the blank-linking line maps, only one proposition was the same across skeleton map C and skeleton map D. Students' responses on each skeleton map were scored as correct or incorrect. A maximum of 12 points could be awarded to each student on each skeleton map.

As in previous studies, to score students' constructed maps we developed a proposition inventory to account for variation in the quality of the students' propositions. This inventory contained the 190 possible relations between a specific pair of concepts in the key-concept list. Based on this inventory, each proposition was scored on a five-point scale, from 0 for inaccurate/incorrect to 4 for excellent/outstanding (complete proposition that showed deep understanding of the relation between two concepts; see Ruiz-Primo, Schultz, & Shavelson, 1996 for a definition of each category). The maximum score for a map constructed by students was based on the criterion map: the number of

links (38) in the criterion map was multiplied by 4 (all propositions were scored as excellent).

After administering the concept maps, all classes received a 30-item multiple-choice test on "Chemical Names and Formulas" designed by the teachers and the researchers. The internal consistency of the test was .74.

Results

In this study we asked the following questions: Are fill-in-the-map scores sensitive to the nodes and the linking lines selected to be filled-in? Are fill-in-the-nodes skeleton maps equivalent to the fill-in-the-linking lines skeleton maps? Does the fill-in-the-map technique provide the same picture of a student's knowledge structure as construct-a-map technique?

Before focusing on these questions, a preliminary issue needs to be addressed, the grouping of students. We initially planned to contrast the advanced and regular chemistry classes. However, when we compared the seven classes using the multiple-choice tests scores, only Class 6 differed significantly from one to two other classes (Classes 2 and/or Class 4). Consequently, we decided to collapse the seven classes and present overall results for simplicity and brevity.

Comparing Concept and Linking Line Sample Scores

To determine whether the fill-in-the-map scores were sensitive to the sample of nodes (concepts) or linking-lines left blank, we compared the mean and variances of scores between Skeleton Map A and B and between Skeleton Map C and D.

The mean scores and standard deviations for the *fill-in-the-nodes* skeleton maps 1 and 2 and *fill-in-the-linking lines* skeleton maps 1 and 2 are presented on Table 1. Overall, students' performance across the two types of skeleton maps and samples was high. However, it was higher for fill-in-the-nodes maps than for fill-in-the-linking lines. Independent-samples t-test indicated no significant difference between the two samples of concepts mean (t=1.57, p=.12) or the two samples of linking lines mean (t = 1.65, p = .10). An F_{Max} test also indicated no significant difference between the variances of the two fill-in-the-nodes (F_{Max}=1.50, p>.05) or the two fill-in-the-linking lines (F_{Max}=1.27,

p>.05) skeleton maps. We concluded that both samples of nodes and linking lines were equivalent and that students' scores were not affected by the particular sample used in the skeleton maps. Similar results were found in one of our previous studies (Ruiz-Primo, Schultz, & Shavelson, 1996).

Table 1

Mean and Standard Deviation by Type of Skeleton Map and Sample

Type of Skeleton Map	n	Mean (max. = 12)	S.D.
Fill-in-the-nodes			
Sample 1	80	11.21	1.42
Sample 2	72	10.80	1.74
Fill-in-the-linking lines			
Sample 1	78	9.77	2.74
Sample 2	73	8.99	3.09

Comparing Types of Fill-In Maps

For the fill-in-the-node and fill-in-the-linking line techniques to be considered equivalent, they need to produce similar means and variances. We carried out a 2 x 4 split-plot ANOVA to evaluate whether the type of skeleton map (i.e., fill-in-the-node and fill-in-the-linking line) and the sequence in which students took the different forms of skeleton maps (e.g., skeleton map A followed by skeleton map C or skeleton map A followed by skeleton map D) affected their scores.

Table 2 provides the mean scores and standard deviations for each type of skeleton map and sequence. ANOVA results indicated a significant interaction between type of skeleton map (T) and sequence (S) (F_{TxS}=2.73, p = .046) and significance difference for type (F_T=65.95, p = .000); no significant difference was observed for sequence (F_S=.63; p = .599).

A closer examination of the interaction showed that it was ordinal. The mean difference in scores between nodes and linking lines skeleton maps was not statistically significant for those students under Sequence 3 (F_{S_3}=3.73, p=.055) whereas it was for those students under the other three sequences (F_{S_1}=13.49, p = .000; F_{S_2}=24.66, p = .000; F_{S_4}=32.53, p = .000). Filling-in-the-nodes using sample 2 for the skeleton map somehow facilitated the fill-in-the-linking

lines when sample 1 was used for the skeleton map. A closer look into the skeleton maps revealed that the number of propositions students needed to read to fill-in-the-nodes in skeleton map B that overlapped with the linking lines they needed to filled-in on skeleton map C was higher than the number observed in any other sequence.

Table 2

Mean and Standard Deviation by Type of Skeleton Map And Sequence

Sequence	Fill-In-The-Node			Fill-In-The-Linking Line	
	n	Mean	S.D.	Mean	S.D.
1 Nodes 1-Lines 1	43	11.09	1.52	9.72	2.84
2 Nodes 1-Lines 2	36	11.03	1.33	9.31	3.06
3 Nodes 2-Lines 1	35	10.63	1.81	9.83	2.65
4 Nodes 2-Lines 2	37	10.97	1.67	8.68	3.13
Total	152	11.02	1.59	9.39	2.93

For the purposes of the study, however, a more important result is the one related to the differences between the types of skeleton maps. The split-plot ANOVA indicated that means differed significantly. Fill-in-the-nodes maps were easier for students than fill-in-the-linking lines maps. Furthermore, the Mauchly's test of sphericity indicated that variances between the two types of skeleton maps also differed significatively (W=1.00, p=.00). Therefore, it was concluded that fill-in-the-node and fill-in-the-linking line are not equivalent forms of skeleton maps.

Since the two samples of nodes and linking lines were considered equivalent and did not have an effect on students' scores, we ignored the sample of nodes or linking lines used in the skeleton maps as well as the sequence and correlate the two types of maps. The correlation between fill-in-the-nodes and fill-in-the-linking lines skeleton maps was .52 (p = .01). The magnitude of the correlation suggests that students were ranked differently across the two types of maps. However, the magnitude of the correlation may be lowered due to the restriction of range observed in the fill-in-the-nodes maps.

Comparing Mapping Techniques

In this section we compare the two mapping techniques, fill-in-the-map and construct-a-map. First, we examine the consistency of scores across raters for the construct-a-map technique. Then we characterize students' constructed maps, and compared the two techniques. Finally, we compare scores from the two mapping techniques with multiple-choice test scores.

Interrater Reliability. All construct-a-maps were scored for accuracy and comprehensiveness. For each student we calculated a: *propositional accuracy* score--the sum of the scores obtained on all propositions; *convergence* score--the proportion of accurate propositions in a student's maps out of all possible propositions in the criterion map; and *salience* score--the proportion of valid proposition out of all the propositions in the student's map.

A sample of 55 students' maps (more than a third of the sample) were scored by three raters. To examine the generalizability of scores across raters, three person (p) by rater (r) G studies were carried out, one for each type of score (Table 3).

Table 3

Estimated Variance Components and Generalizability Coefficients for a Person by Rater G Study Across Types of Scores

	Score Type					
	Proposition Accuracy		Convergence		Salience	
Source of Variation	Estimated Variance Component	Percent of Total Variability	Estimated Variance Component	Percent of Total Variability	Estimated Variance Component	Percent of Total Variability
Persons (p)	290.54	96.26	0.03114	97.65	0.02863	95.15
Raters (r)	0.36	0.12	0.00011	0.34	0.00020	.66
pr,e	10.92	3.62	0.00064	2.00	0.00126	4.19
$\hat{\rho}^2$	.99		.99		.98	
$\hat{\phi}$	.99		.99		.98	

Results indicated that the error introduced by raters was negligible. Both relative and absolute coefficients were very high across types of scores. Based on these results, the remaining 97 concept maps were randomly distributed

among the three raters and only one rater scored each map. The randomization was done within each of the seven classes. Thus, all three raters scored a sample of students' maps across the seven classes.

Students' Maps. Table 4 provides information about the characteristics of students' constructed maps. Two thirds of the maps used all 20 concepts provided in the list to construct their maps. Another fifth used 18-19 concepts and only one student used just 14 concepts only.

Table 4

Means and Standard Deviations of Students' Concept Map Components

Map Components	n	Mean	S.D.	Min	Max
Nodes in the Map	152	19.34	1.23	14	20
Linking Lines	152	25.41	6.60	14	43
Accurate Propositions	152	18.88	7.44	0	42

A surprising finding was that 6.6 percent of the students provided more than 38 links in their maps, which is the number of links on the criterion map.[3] Furthermore, 40 percent of these students provided more than 38 accurate propositions.

It is important to mention that a few of the students provided better propositions than those in the criterion map! This led us to re-score the criterion map using the same criteria applied for students. Therefore, some propositions in the criterion map became "Good," instead of "Excellent," and one proposition became "Poor." The original maximum was 158 and was corrected to 135.

Students' Scores Across Assessment Techniques. Table 5 provides the descriptive statistics for the three types of assessments administered to the students: construct-a-map, fill-in-the-map, and multiple-choice test.

Mean scores across the forms of assessments do not provide the same picture about students' knowledge of the topic. Whereas salience, fill-in-the-map and multiple-choice scores indicate that students' performance was close to the maximum criterion, the proposition accuracy and convergence scores

[3] In fact, 18 percent of students provided between 25 and 38 links.

indicate that students' knowledge was rather partial compared to the criterion map.

Table 5

Means and Standard Deviations Across the Three Types of Assessments Administered to Students

	n	max.	Mean	S.D.
Construct-A-Map				
Proposition Accuracy	152	135	53.91	22.17
Convergence	152	1	.50	.19
Salience	152	1	.73	.17
Fill-In				
Fill-In-The-Nodes	152	12	11.02	1.59
Fill-In-The-Linking Lines	151	12	9.39	2.93
Multiple-Choice Test	150	30	24.05	3.74

All types of scores, except proposition accuracy and convergence, showed negatively skewed distributions (skewness value ranged from -.755 for fill-in-the-linking lines, to -1.538 for fill-in-the-nodes) indicating that most of the students obtained high scores. Furthermore, the Kolmogorov-Smirnov normality test confirmed that only proposition accuracy and convergence scores were normally distributed (p = .200). It seems that proposition accuracy and convergence scores better reflect the differences in students' knowledge than the other scores.

A correlational approach was used to compare techniques because of the different score scales across techniques. Table 6 provides a multiscore-multitechnique matrix. We first focus on comparing scores within each mapping technique. Then, we evaluate the extent to which the scores on the two mapping techniques converge, and finally, we evaluate the extent to which the two mapping technique scores converge with multiple-choice scores.

In the matrix, reliability coefficients are enclosed in parenthesis on the diagonal. Along with the observed correlations, we present correlations corrected for unreliability.[4] However, because different reliability estimates

[4] No correction needed for the construct-a-map technique.

are used in the matrix, and hence error measurement is defined differently, some of these corrections may not be accurate and must be interpreted cautiously. Therefore, we focus on the observed correlations.

Table 6

Multiscore-Multitechnique Matrix

	Construct-A-Map			Fill-In-The-Map		MC
	PA	CON	SAL	NOD	LIN	
Construct-A-Map						
Proposition-Accuracy (PA)	(.99)[a]					
Convergence (CON)	.95	(.99)[a]				
Salience (SAL)	.73	.75	(.98)[a]			
Fill-In-The-Map						
Fill-in-the-nodes (NOD)						
Observed	.50	.47	.45	(.70)[b]		
Corrected	.61	.56	.54			
Fill-in-the-lines (LIN)						
Observed	.51	.44	.40	.53	(.84)[b]	
Corrected	.56	.49	.44	.69		
Multiple-Choice (MC)						
Observed	.51	.44	.46	.37	.65	(.74)[c]
Corrected	.60	.51	.54	.51	.83	

[a] Interrater reliability.
[b] Internal consistency averaged between the two skeleton maps.
[c] Internal consistency.

Construct-A-Map Scores. The correlation between proposition accuracy and convergence scores is very similar to correlations we have found in other studies (e.g., Ruiz-Primo, Schultz, & Shavelson, 1996; Ruiz-Primo, Shavelson, & Schultz, 1997). This very high correlation suggests that both scores rank students similarly. Furthermore, when G theory has been used to evaluate the dependability of these measures (see Ruiz-Primo, Schultz, & Shavelson, 1996; Ruiz-Primo, Shavelson, & Schultz, 1997), we found that the percent of variability among persons is higher for proposition accuracy and convergence scores than for salience scores. This indicates that these two measures better reflect the differences in students' knowledge structures than do salience scores.

The correlations between proposition accuracy and convergence scores with salience scores (.73 and .75 respectively), however, are lower than the ones we have observed before (~.85). A possible reason for this lower

correlation may be students' knowledge level. In this study, students clearly had better knowledge about the topic than students we tested before. The means obtained in this study were impressively higher when compared with those we obtained before (Proposition Accuracy = ~11; Convergence = ~.17; Salience = ~.50). Students in this study provided more accurate propositions in their maps, thereby improving their salience scores. In our previous studies students' scores were low across types of scores so their ranking did not differ across scores.

The general conclusion about construct-a-map scores is consistent with our previous research. Proposition accuracy and convergence scores reflect the differences in students' knowledge structure better than do salience scores. Based on practical (e.g., scoring time) and technical (e.g., instability of scores) arguments, we conclude that the convergence score is the most efficient.

Mapping Technique Scores. If the construct-a-map and fill-in techniques measure the same construct, we should expect a high correlation among these scores. Yet, correlations were lower than expected (r=.46 averaged across types of scores). Restriction of range observed in both types of fill-in-the-map scores may contribute to the magnitude of the correlations; interpretation of the low coefficients should be considered with caution.

Although correlations between fill-in-the-nodes and fill-in-the-linking lines with construct-a-map scores were not the same magnitude (correlations are higher between fill-in-the-nodes and construct-a-map), no significant difference (p >.05) was found among the correlations ($\chi^2_{(5)}$ = 3.23; Meng, Rosenthal & Rubin, 1992). The pattern of correlations, however, is the same: The highest correlation is with proposition accuracy and the lowest with salience scores. The magnitude of the correlations of both series of coefficients indicate that students are ranked differently according to the technique used. It seems that different aspects of the students' connected understanding are being tapped with the construct-a-map technique and the fill-in-the-map technique.

Comparing Mapping and Multiple-Choice Scores. Correlations between multiple-choice tests scores and each type of construct-a-map scores are similar to the ones observed between the fill-in-the-map and construc-a-map scores. No significant difference was observed among the nine correlations

($\chi^2_{(8)} = .161$; $p > .05$). We concluded that construct-a-map scores correlated similarly with fill-in-the-map and multiple-choice scores.

The correlations between fill-in-the-map scores with multiple-choice scores were quite surprising. The magnitude of the correlations between fill-in-the-nodes and multiple-choice test reported by Schau et al. (1996) are higher (r=.75 on average) than the one we found in this study (r=.37).

Two issues may explain these differences: restriction of range observed in the fill-in-the-nodes skeleton map scores (i.e., skeleton map was very easy for students in our study) and differences between the characteristics of the fill-in-the-nodes maps used in both studies (e.g., Schau et al., 1997, used 37 nodes; 50 percent were left blank; we used 20, and 60 percent were left blank. Also, the propositions in the skeleton map used by Schau et al. were less complex than the ones used in ours). Whether the characteristics of the maps can affect students' scores is a topic that deserves to be studied more carefully. For example, what number of nodes in a skeleton map are optimum? How many nodes need to be left blank? What is the best way to select the nodes left blank?

The correlation between fill-in-the-linking lines is the highest among all the correlations between mapping scores with multiple-choice tests. Both score distributions share about 43 percent of the variance, whereas only 14 percent is shared with the fill-in-the-nodes scores. There may be greater similarity between multiple-choice test and the fill-in-the-linking lines map than with fill-in-the-nodes map. Differences between these two forms of fill-in maps deserve more attention.

An important finding for our purposes is that the pattern of correlations is not same across mapping techniques. The pattern is more similar within the construct-a-map scores, than within the fill-in-the-map scores. Mapping techniques, then, do not provide similar information about students' knowledge structure or connected understanding.

We think that the construct-a-map technique better reflects the state of students' knowledge structure. We based this conclusion on the fact that this technique is the only one which accurately reflects the differences among students' scores. But, what is the fill-in-the-map technique tapping? What aspect of the students' knowledge is being measured with this form of assessment? A closer look at the cognitive activities displayed in this technique

is needed. Talk aloud protocols may help to better define the cognitive activities reflected by both techniques.

An overall conclusion is that we need to invest time and resources in finding out more about what aspects of students' knowledge are tapped by different forms of assessment. What makes those assessments share variance? What is the unique variance? Which technique should be considered the most appropriate for large-scale assessment? Practical issues, though, cannot be the only criterion for selection. Students' partial knowledge may be hidden more easily on some form of assessment than on others. To resolve the issue of what is being measured with these different techniques, we need information about the cognitive activity displayed on each of them.

Conclusion

In this study we explored the equivalence of two mapping techniques, fill-in-the-map and construct-a-map from scratch. We examined whether: (1) skeleton map scores were sensitive to the sample of nodes or linking lines left blank, (2) the two forms of skeleton maps were equivalent, and (3) the two mapping techniques provided similar information about students' connected understanding.

Our results led to the following tentative conclusions: (1) skeleton map scores are not sensitive to the sample of concepts or linking lines to be filled-in. Probably the list of concepts and propositions was cohesive enough so that any combination of concepts or propositions could provide similar information about students' knowledge. (2) Fill-in-the-nodes and fill-in-the-linking lines techniques are not equivalent forms of fill-in-the-map. Further research is needed to define which of these two forms provide the most accurate information about students' knowledge or connected understanding. (3) The relationship between the two mapping techniques studied suggests that both mapping techniques are tapping somewhat but not identical aspects of students' connected understanding. As we previously suggested, talk aloud protocols may provide insight about the cognitive activities involved in constructing and filling-in a map. (4) Construct-a-map scores most accurately reflected the differences across students' knowledge structure. (5) The relationship between scores from the multiple-choice test and both mapping techniques confirmed that mapping techniques were not equivalent. The

pattern of correlation coefficients was different across mapping techniques. (6) Convergence scores--the proportion of accurate propositions in the students' maps to the number of all possible propositions in the criterion map--is the most efficient indicator when scoring construct-a-map concept maps.

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